Epistemic Uncertainty in the Prediction of Engineering Systems

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Outline of Presentation



- Programmatic background
- Epistemic uncertainty project
- Epistemic uncertainty in computational modeling
- Example problem
- Project activities
- Status and future work

Programmatic Background



- ASCI Verification and Validation Program Element:
 - Program manager: Marty Pilch, Org. 9133
 - Technical lead: Tim Trucano, Org. 9211
- Major tasks in the V&V program:
 - Planning and assessment of verification, validation, and software quality assurance practices
 - Validation metrics
 - Validation simulations for stockpile-to-target-simulations (normal, abnormal, and hostile environments)
 - Uncertainty quantification

ASCI Epistemic Uncertainty Projects



- Non-distributional probabilistic approach:
 - John Red-Horse, PI
 - Relaxation of full specification of probability distributions
- Epistemic uncertainty project:
 - Bill Oberkampf, PI
 - Generalization of traditional probability theory
- Probabilistic robustness analysis:
 - Steve Wojtkiewicz, Pl
 - Propagation of interval and random variable uncertainty

Epistemic Uncertainty Project



- Began in Q3 of FY00
- Team Members:
 - Bill Oberkampf (fluid dynamics and heat transfer)
 - Jon Helton (traditional probabilistic risk assessment)
 - Cliff Joslyn (generalized information theory)
 - Scott Ferson (non-traditional probabilistic risk assessment)
 - Steve Wojtkiewicz (traditional probabilistic dynamics)
 - Kari Sentz (generalized information theory)
- Project goal:
 - Evaluate feasibility of generalizations of probability theory for uncertainty estimation in large-scale simulations.

Epistemic Uncertainty Project Approach



- Methods must be directly compatible with traditional probabilistic approaches.
- Emphasis is on improving uncertainty estimation when data is scarce.
- Methods must be usable with existing large-scale computational codes, i.e., "black box" codes
- Improvement of non-traditional methods over traditional methods will be judged based on comparison of:
 - Ability to construct uncertainty representations of input data
 - Fidelity of mapping input uncertainty to output uncertainty
 - Required number of computational model function evaluations
 - Ability to conduct sensitivity analyses

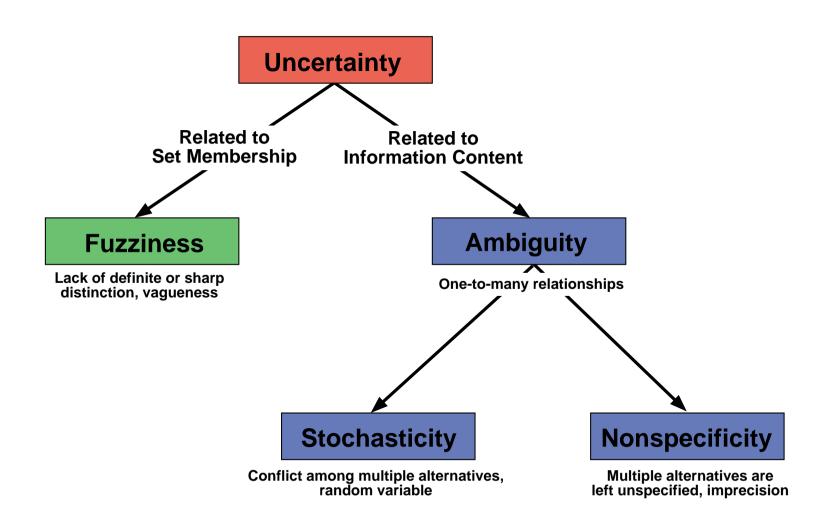
Epistemic Uncertainty in Computational Modeling



- Parametric uncertainty:
 - Uncertainty characterized by a probability density function with interval valued parameters
 - Expert opinion input characterized by interval values and varying levels of belief over an interval
- Mathematical modeling uncertainty:
 - Uncertainty in the specification of an event tree or fault tree
 - Estimate uncertainty using alternate plausible mathematical models
- Scenario abstraction uncertainty:
 - Uncertainty in the identification of event or fault sequences
 - Uncertainty in the likelihood of the event or fault sequences

Generalized Types of Uncertainty





Evidence Theory as a Generalization of Probability Theory



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- Non-additive measures of evidence theory:
 - Belief measure:

$$Bel(A) = \sum_{B \subseteq A} m(B)$$

Plausibility measure

$$Pl(A) = \sum_{B \cap A \neq \emptyset} m(B)$$

 Belief and plausibility measures can be thought of as lower and upper probabilities, respectively:

$$Bel(A) \leq Pl(A)$$

Example Problem

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• Let the mathematical model for a physical process be given by

$$y = a^b$$

 Assume the only information concerning a and b is given by the nested sets:

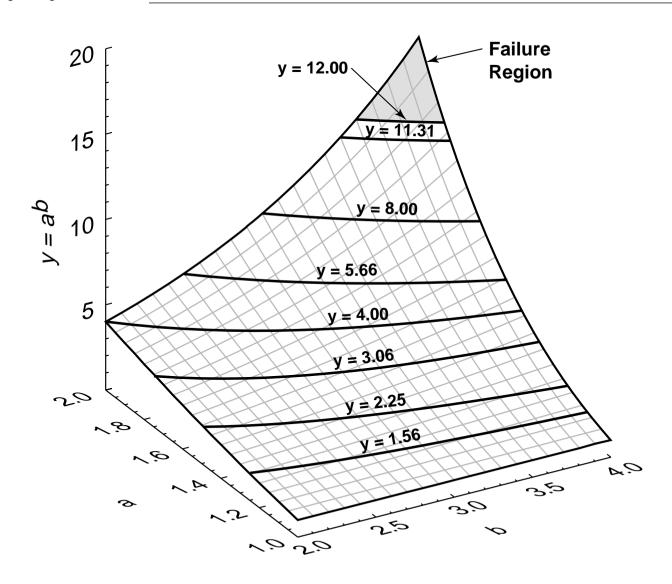
•
$$A_1 = [1, 1.25], A_2 = [1, 1.5], A_3 = [1, 1.75], and A_4 = [1, 2]$$

•
$$B_1 = [2.5,3], B_2 = [2.5,3.5], B_3 = [2,3.5], and $B_4 = [2,4]$$$

- Assume the system fails when y > 12
- Goal of the analysis:
 - Given the data for a and b, what can be said about the occurrence of y > 12?

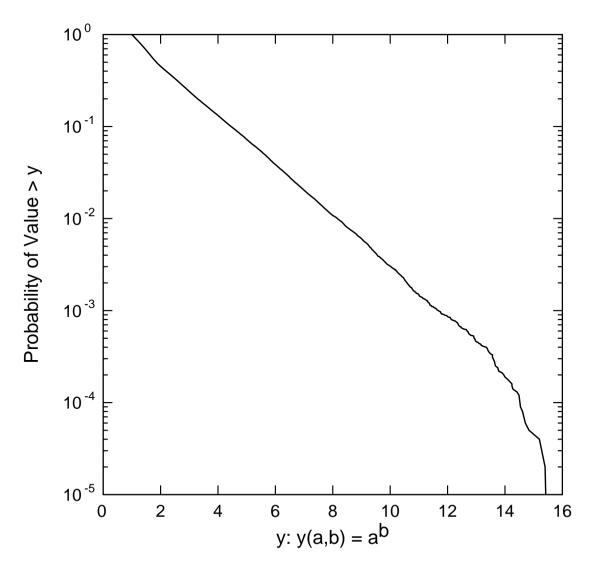
Character of the System Response





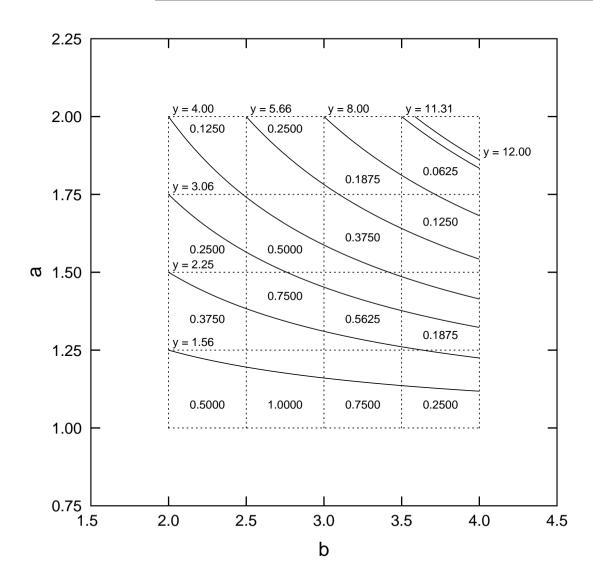
Solution from Traditional Probability Theory: Complementary CDF





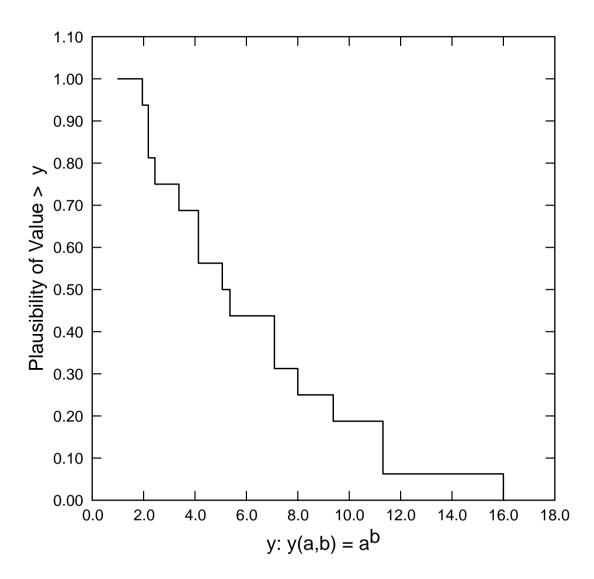
Plausibility Values for the Product Space AXB





Solution with Evidence Theory: Complementary CPF





Project Activities



- Constructed a set of "Challenge Problems":
 - Epistemic uncertainty exists only in input parameters for two simple computational models
 - Varying levels of epistemic uncertainty, from only interval values to specified PDFs with interval valued parameters
 - Two simple systems, algebraic system and an initial value problem given by a linear ordinary differential equation
 - Solutions from different approaches will be compared
- Constructed a web site describing the project and the challenge problems:
 - www.sandia.gov/epistemic/

Workshop on Epistemic Uncertainty



- Workshop will be held on August 6-7, 2002 in Albuquerque.
- Will have invited papers and submitted papers presenting solutions to one or more of the Challenge Problems.
- Papers can also pursue related topics in epistemic uncertainty
- Invited speakers cover a wide range of approaches:
 - Frequentist
 - Subjectivist
 - Information theory
 - Risk analysts
- Invited and submitted papers will be reviewed for publication in a special issue of *Reliability Engineering and System Safety*.

Status and Future Work



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Present work:

- Methods for construction of uncertainty representations from a variety of sources with emphasis on scarce data
- Survey of methods for aggregation of data with varying degrees of conflict
- Proofs of convergence for Belief and Plausibility mapping of input to output
- Improved understanding of factors affecting rate of convergence of Belief and Plausibility of input to output

Future work:

- Improved understanding of the relationship of probabilitybounds to evidence theory
- Development of convergence acceleration techniques for Belief and Plausibility